

The problem of predecessors on spanning trees

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We consider the equiprobable distribution of spanning trees on the square lattice. All bonds of each tree can be oriented uniquely with respect to an arbitrary chosen site called the root. The problem of predecessors is finding the probability that a path along the oriented bonds passes sequentially fixed sites i and j . The conformal field theory for the Potts model predicts the fractal dimension of the path to be $5/4$. Using this result, we show that the probability in the predecessors problem for two sites separated by large distance r decreases as $P(r) \sim r^{-3/4}$. If sites i and j are nearest neighbors on the square lattice, the probability $P(1) = 5/16$ can be found from the analytical theory developed for the sandpile model. The known equivalence between the loop erased random walk (LERW) and the directed path on the spanning tree says that $P(1)$ is the probability for the LERW started at i to reach the neighboring site j . By analogy with the self-avoiding walk, $P(1)$ can be called the return probability. Extensive Monte-Carlo simulations confirm the theoretical predictions.

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I. INTRODUCTION AND MAIN RESULTS

In the graph theory, the spanning tree of connected graph G is a connected subgraph of G containing all vertices of G and having no cycles. Numerous applications of spanning trees began with the seminal Kirchhoff's problem solved in 1847 and then spread out many branches of mathematics and theoretical physics. In the statistical mechanics, spanning trees are related to the Potts model [1], the dimer model [2], the sandpile model [3] and many others. A relation between lattice models of statistical mechanics and spanning trees via the Tutte polynomial has been established by Fortuin and Kasteleyn [4].

The Kirchhoff theorem claims that the number of spanning trees of connected graph G is a cofactor of the Laplacian matrix Δ of graph G . If one deletes any row and any column from Δ , one obtains a matrix Δ^* which gives the number of spanning trees as $\det \Delta^*$. The determinantal structure allows easy calculation of local characteristics of the spanning trees, for instance, the average number of vertices with given number of adjacent bonds. A characterization of non-local objects in the spanning tree is not so simple. One of such the objects is the chemical path defined

as a path along two or more bonds of the tree. The fractal dimension of a long chemical path on the two-dimensional lattice has been calculated by means of a mapping of the spanning tree configurations onto the Coulomb gas model.

A closely related object is the loop erased random walk (LERW) on the two-dimensional lattice [11] which was proven to be equivalent to the directed chemical path of the spanning tree of the same lattice [7, 12]. In this paper, we consider a problem arising in the theory of LERW and equally distributed spanning trees: given two lattice sites i and j , what is the probability that the LERW or the directed chemical path passes i and j . If site i is passed first, we say that i is the predecessor of j and coin the mentioned problem as the predecessor problem. Surprisingly, the problem has no exact solution in the general case. Only two limiting cases are available: (a) If sites i and j are separated by large distance r , the asymptotics of $P(r)$ can be found from known results on the fractal dimension of the chemical path; (b) If points i and j are nearest neighbors of the square lattice, the seeking probability can be found from the theory of sandpiles [9] (see also [16]).

The asymptotic behavior of $P(r)$ for large distance r follows directly from the definition of fractal dimension. Indeed, consider a large square lattice \mathcal{L} and the set of uniformly distributed spanning trees on \mathcal{L} . We assume that the root is situated at the boundary of \mathcal{L} . Consider site i in the bulk of the lattice and some circle contour C of radius R with the center in i . Let Π be a directed chemical path from i to the root along the oriented bonds of a tree. All points of the subset of Π inside C are descendants of i . In accordance with the definition of the fractal dimension of the directed path on the spanning tree, the number of the descendants inside C is proportional to $R^{5/4}$ (see Majumdar [7]). The probability that point i is the predecessor of point j lying at distance r from i is the density of descendants $\rho(r)$. Thus, we have

$$\int_1^R \rho(r) r dr \sim R^{5/4} \quad (1.1)$$

from where we conclude that $\rho(R) \sim R^{-3/4}$.

As it was mentioned before, the problem of predecessors for an arbitrary disposition of two lattice points is not solved. In Section 2 we concentrate on a particular problem of probability $P(1)$ when points i and j are nearest neighbors of the square lattice. An essential element of the theory of sandpiles is the probability distribution of sites having 0, 1, 2 and 3 predecessors among the nearest neighbors. The corresponding probabilities are denoted by X_0 , X_1 , X_2 and X_3 . Having explicit expressions for these values, we obtain $P(1)$ as their combination and get an unexpectedly simple result $P(1) = 5/16$. In Section 3 we relate this result to the return probability of the LERW. Section 4 contains results of the Monte-Carlo verifications.

II. THE PROBLEM OF PREDECESSORS FOR NEAREST NEIGHBORS

The spanning tree enumeration method, namely, the Kirchhoff theorem is proved to be a powerful mathematical instrument for the investigation of various combinatorial problems of the theoretical physics. In the last decade, it has been developed and adapted for the calculation of height probabilities of the Abelian sandpile model [5, 9, 16]. The Abelian sandpile model is a stochastic dynamical system, which describes the phenomenon of self-organized criticality. During the evolution the system falls into a subset of all possible states, called the subset of recurrent states. The problem is to calculate analytically various observable values in the recurrent state, such as height probabilities P_i , $i = 1, 2, 3, 4$ at a fixed site and height correlations between distinct fixed sites [18].

It was shown that the calculation of height probabilities in the Abelian sandpile can be reduced to the calculation of X_0 , X_1 , X_2 and X_3 in the spanning tree model. The exact relation between these quantities is given by

$$P_1 = \frac{X_0}{4}; \quad P_2 = P_1 + \frac{X_1}{3}; \quad P_3 = P_2 + \frac{X_2}{2}; \quad P_4 = P_3 + X_3. \quad (2.1)$$

Majumdar and Dhar [5] have found in 1991 the probability of height 1, constructing the corresponding defect lattice for the situation when a site i_0 has no predecessors and calculating the determinant of the defect matrix Δ^* . A technique for computing the numbers X_1, X_2, X_3 has been devised in [9]. The results are (see also [16] for details)

$$X_0 = \frac{8(\pi - 2)}{\pi^3}; \quad X_1 = \frac{3}{4} + \frac{48}{\pi^3} - \frac{15}{\pi^2} - \frac{3}{2\pi}; \quad X_2 = \frac{1}{4} - \frac{48}{\pi^3} + \frac{6}{\pi^2} + \frac{3}{\pi}; \quad X_3 = \frac{16}{\pi^3} + \frac{1}{\pi^2} - \frac{3}{2\pi}, \quad (2.2)$$

which give

$$P_1 = \frac{2(\pi - 2)}{\pi^3}; \quad P_2 = \frac{1}{4} + \frac{12}{\pi^3} - \frac{3}{\pi^2} - \frac{1}{2\pi}; \quad P_3 = \frac{3}{8} - \frac{12}{\pi^3} + \frac{1}{\pi}; \quad P_4 = \frac{3}{8} + \frac{4}{\pi^3} + \frac{1}{\pi^2} - \frac{1}{2\pi}. \quad (2.3)$$

Now consider the problem of predecessors for nearest neighboring sites. First we fix a site i_0 in the bulk of the square lattice. Denote its right nearest neighboring site by j_0 . Next consider four various cases, when i_0 has exactly k nearest neighboring predecessors ($k = 0, 1, 2, 3$) (see Fig 1). For $k = 0$ the site j_0 trivially is not a predecessor of i_0 . For $k = 1$, we have 1 of 4 equivalent situations when j_0 is the predecessor of i_0 . For $k = 3$, we have 3 of 4 equivalent situations when j_0 is the predecessor. The crucial case is the $k = 2$. Here we have 6 situations, but not all of them are equivalent. On the other hand, we can select two groups of 4 and 2 elements (the first four and the last two in the third line of Fig. 1) so that elements in each group are equivalent. We are looking for the situations where j_0 is a predecessor of i_0 . There are 2 encircled elements from the first group and one from the second group, which correspond to the desired situations. Thus, if we take the linear combination of X_1 , X_2 and X_3 with coefficients $1/4$, $1/2$ and $3/4$ correspondingly,

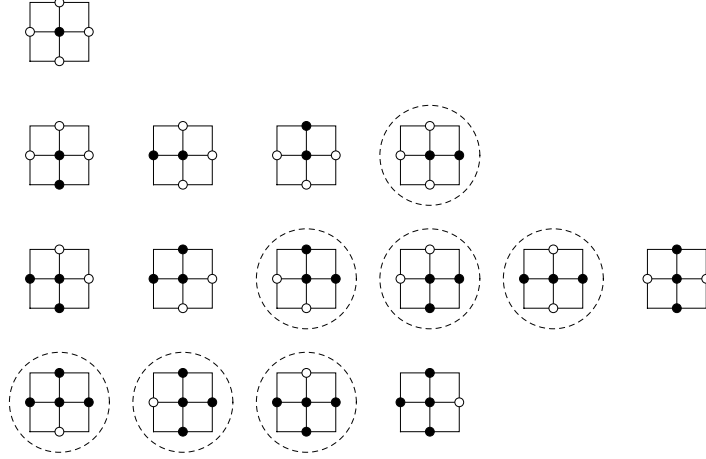


FIG. 1: All possible situations with a fixed vertex i_0 (the central vertex in diagrams) to have various nearest neighbouring predecessors on square lattice. By black color we denote vertices, which are predecessors of i_0 . White means that the corresponding vertex is not a predecessor of i_0 . On the $k + 1$ st row ($k = 0, 1, 2, 3$) situations with k predecessors are presented. By the dashed line are circled the configurations for which the right neighboring vertex is a predecessor of i_0 .

we get the desired probability $P(1)$ that j_0 is the predecessor of i_0 :

$$P(1) = \frac{1}{4}X_1 + \frac{1}{2}X_2 + \frac{3}{4}X_3 = \frac{5}{16}. \quad (2.4)$$

III. RETURN PROBABILITY FOR THE LOOP ERASED RANDOM WALK

Consider a finite square lattice \mathcal{L} with vertex set V and edge set E . Given $\mathcal{P} = [u_0, u_1, u_2, \dots, u_n]$ a path in \mathcal{L} , its loop-erasure $\text{LE}(\mathcal{P}) = [\gamma_0, \gamma_1, \gamma_2, \dots, \gamma_m]$ is defined by chronologically removing loops from \mathcal{P} . Formally, this definition is as follows. We first set $\gamma_0 = u_0$. Assuming $\gamma_0, \dots, \gamma_k$ have been defined, we let $s_k = 1 + \max\{j : u_j = \gamma_k\}$. If $s_k = n + 1$, we stop and $\text{LE}(\mathcal{P}) = [\gamma_0, \gamma_1, \gamma_2, \dots, \gamma_m]$ with $m = k$. Otherwise, we let $\gamma_{k+1} = u_{s_k}$. Note that the order in which we remove loops does matter, and it follows from the definition, we remove loops as they are created, following the path. A walk, obtained after applying the loop-erasure to a simple random walk path is called Loop-Erased Random Walk (LERW). Since the infinite simple random walk on finite connected undirected graphs is recurrent, the infinite LERW is not defined. On the other hand, we can fix a subset $W \subset V$ of vertices and define LERW from a fixed vertex u_0 to W . To do that we take a path of a simple random walk started at u_0 and stopped upon hitting W , after that we apply loop-erasure.

Wilson established an algorithm to generate uniform spanning trees, which uses LERW [13]. It turns out to be extremely useful not only as a simulation tool, but also for theoretical analysis. It runs as follows. Pick an arbitrary ordering $V = \{v_0, v_1, \dots, v_N\}$ for the vertices in \mathcal{L} . Let $S_0 = v_0$.

Inductively, for $i = 1, 2, \dots, N$ define a graph S_i to be the union of S_{i-1} and a (conditionally independent) LERW path from v_i to S_{i-1} . Note, if $v_i \in S_{i-1}$, then $S_i = S_{i-1}$. Then, regardless of the chosen order of the vertices, S_N is a UST on \mathcal{L} with root v_0 . If we take as an initial condition $S_0 = W$, with some $W \subset V$, then the generated structures will be spanning forests with set of roots W . The spanning forest with fixed set of roots can be considered as a spanning tree, if we add an auxiliary vertex and join it to all the roots.

Now consider the Wilson algorithm on \mathcal{L} with the set of boundary vertices $\partial\mathcal{L}$ and take $S_0 = \partial\mathcal{L}$. When the size of the lattice tends to infinity, the boundary effects will vanish, so we can neglect the details of the boundary. So we will not distinguish between spanning forests and spanning trees. It follows from the Wilson algorithm for a fixed site i_0 , that if we start a LERW from i_0 upon hitting the boundary $\partial\mathcal{L}$, we will generate a path ℓ of a spanning tree from i_0 to the boundary (see also [7], [8]). All vertices on the path ℓ form the set of descendants of i_0 . So, if a fixed vertex j_0 belongs ℓ , i_0 is a predecessor of j_0 .

Consequently, the probability $P(j_0 - i_0)$ that i_0 is a predecessor of j_0 in a randomly taken (from the uniform distribution) spanning tree on the large square lattice equals to the probability that the LERW started from i_0 passes j_0 . In the particular case $|j_0 - i_0| = 1$, the probability $P(1) = 5/16$ calculated in the previous section is the return probability for the LERW.

IV. MONTE-CARLO SIMULATIONS

Consider finite $2N + 1 \times 2N + 1$ square lattice \mathcal{L}_N . Denote its central vertex by i_0 and assume that it is an origin of the coordinate system. We deliberately took an odd-odd lattice to provide the symmetry which guarantee the most efficient vanishing of boundary effects for large lattices. After generating a large amount of LERWs starting from i_0 , we get an approximation of the function $P_N(j_0 - i_0) \equiv P_N(j_0)$. Given fixed j_0 , we can extrapolate $P_N(j_0)$, tending N to infinity and get asymptotical function $P(j_0)$. Assume that the Euclidean distance between the origin i_0 and j_0 is r and coordinates of j_0 are $(r \cos \varphi, r \sin \varphi)$. The Monte-Carlo simulations and Coulomb gas arguments show that the asymptotical behaviour of the function $P(j_0)$ for large r ($r \gg 1$) does not depend on φ . So, for $r \gg 1$ we have $P(j_0) \simeq P(r)$. Fig. 2 shows the behaviour of $P(r)$ for various j_0 on the log-log scale, obtained from Monte-Carlo simulations. From this result we conclude that $P(r)$ decreases as a power function:

$$P(r) \simeq \frac{C}{r^\alpha}, \quad (4.1)$$

with $\alpha \approx 0.751$ and $C \approx 0.305$. During the simulations, we took sizes up to $N = 100$ and number of simulations 10^8 . The obtained results are in agreement with $\alpha = 3/4$ which follows from the Coulomb gas arguments mentioned above. The effective Monte-Carlo algorithm allows evaluation

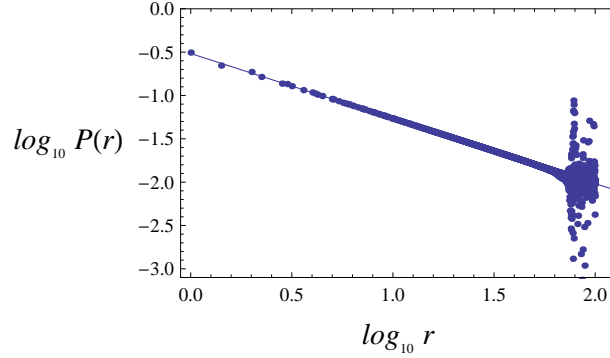


FIG. 2: The results of Monte-Carlo simulations for the probability $P(r)$.

of probabilities $P(j_0 - i_0)$ for arbitrary finite j_0, i_0 . At the same time, the analytical calculation of $P(r)$ for $r > 1$ remains a difficult unsolved problem.

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